```
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
from scipy import stats
from keras.datasets import imdb
from keras.preprocessing.sequence import pad_sequences
from keras.models import Sequential
from keras.layers import Embedding #changed from keras.layers.embeddings to keras.layers
from keras.layers import SimpleRNN, Dense, Activation
\# Input data files are available in the read-only "../input/" directory
# For example, running this (by clicking run or pressing Shift+Enter) will list all files under the input directory
import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))
(X_train,Y_train),(X_test,Y_test) = imdb.load_data(path="imdb.npz",num_words=None,skip_top=0,maxlen=None,
                                                     start_char=1,seed=13,oov_char=2,index_from=3)
    Downloading data from <a href="https://storage.googleapis.com/tensorflow/tf-keras-datasets/imdb.npz">https://storage.googleapis.com/tensorflow/tf-keras-datasets/imdb.npz</a>
     17464789/17464789 -
                                            – 0s 0us/step
print("Type: ", type(X_train))
print("Type: ", type(Y_train))
Type: <class 'numpy.ndarray'>
Type: <class 'numpy.ndarray'>
print("X train shape: ",X_train.shape)
print("Y train shape: ",Y_train.shape)
→ X train shape: (25000,)
     Y train shape: (25000,)
print(X_train[0])
(1, 608, 13, 6467, 14, 22, 13, 80, 1109, 14, 20, 584, 18, 231, 72, 141, 6, 783, 254, 189, 7060, 13, 100, 115, 106, 14, 20, 584, 207,
    4
review_len_train = []
review_len_test = []
for i,j in zip(X_train,X_test):
    review_len_train.append(len(i))
    review_len_test.append(len(j))
print("min: ", min(review_len_train), "max: ", max(review_len_train))
→ min: 11 max: 2494
print("min: ", min(review_len_test), "max: ", max(review_len_test))
→ min: 7 max: 2315
sns.distplot(review_len_train,hist_kws={"alpha":0.3})
sns.distplot(review_len_test,hist_kws={"alpha":0.3})
```

<ipython-input-9-7037aabe6f55>:1: UserWarning: `distplot` is a deprecated function and will be removed in seaborn v0.14.0. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms). For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751 sns.distplot(review_len_train,hist_kws={"alpha":0.3}) <ipython-input-9-7037aabe6f55>:2: UserWarning: `distplot` is a deprecated function and will be removed in seaborn v0.14.0. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms). For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751 sns.distplot(review_len_test,hist_kws={"alpha":0.3}) <Axes: ylabel='Density'> 0.005 0.004 0.003 0.002 0.001 0.000 0 500 1000 1500 2000 2500 4 print("Train mean: ",np.mean(review_len_train)) print("Train median: ",np.median(review_len_train)) print("Train mode: ",stats.mode(review_len_train)) Train mean: 238.71364 Train median: 178.0 Train mode: ModeResult(mode=132, count=196) # number or words word_index = imdb.get_word_index() print(type(word_index)) 5 Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/imdb word index.json 1641221/1641221 <class 'dict'> def whatItSay(index=24): reverse_index = dict([(value,key) for (key,value) in word_index.items()]) decode_review = " ".join([reverse_index.get(i-3, "!") for i in X_train[index]]) print(decode_review) print(Y_train[index]) return decode_review decoded_review = whatItSay()

decoded_review = whatItSay()

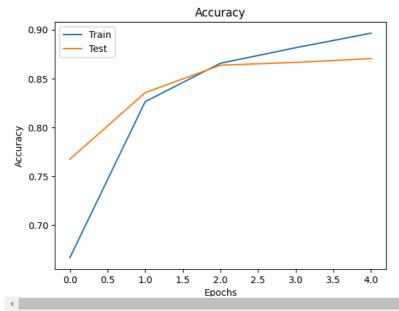
! this movie was extremely funny i would like to own this for my vintage collection of 1970s movie must see again list i know this of the second of the least when sticking to his native france managed.

! quite possibly how francis veber one of the best comedy directors in the world at least when sticking to his native france managed.

ु ! quite possibly how francis veber one of the best comedy directors in the world at least when sticking to his native france managed

```
num\_words = 15000
(X_train,Y_train),(X_test,Y_test) = imdb.load_data(num_words=num_words)
maxlen=130
X_train = pad_sequences(X_train, maxlen=maxlen)
X_test = pad_sequences(X_test, maxlen=maxlen)
print(X train[5])
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rnn = Sequential()
rnn.add(Embedding(num_words,32,input_length =len(X_train[0]))) # num_words=15000
rnn.add(SimpleRNN(16,input_shape = (num_words,maxlen), return_sequences=False,activation="relu"))
rnn.add(Dense(1)) #flatten
rnn.add(Activation("sigmoid")) #using sigmoid for binary classification
print(rnn.summary())
rnn.compile(loss="binary_crossentropy",optimizer="rmsprop",metrics=["accuracy"])
🚁 /usr/local/lib/python3.10/dist-packages/keras/src/layers/core/embedding.py:90: UserWarning: Argument `input_length` is deprecated. I
       warnings.warn(
     /usr/local/lib/python3.10/dist-packages/keras/src/layers/rnn/rnn.py:204: UserWarning: Do not pass an `input_shape`/`input_dim` argum
     super().__init__(**kwargs)
Model: "sequential"
                                               Output Shape
       Layer (type)
                                                                                      Param #
       embedding (Embedding)
                                               ?
                                                                                  0 (unbuilt)
                                               ?
       simple rnn (SimpleRNN)
                                                                                  0 (unbuilt)
       dense (Dense)
                                               ?
                                                                                  0 (unbuilt)
                                               ?
       activation (Activation)
                                                                                  0 (unbuilt)
      Total params: 0 (0.00 B)
      Trainable params: 0 (0.00 B)
      Non-trainable params: 0 (0.00 B)
     None
history = rnn.fit(X_train,Y_train,validation_data = (X_test,Y_test),epochs = 5,batch_size=128,verbose = 1)
⇒ Epoch 1/5
     196/196
                                — 13s 51ms/step - accuracy: 0.5938 - loss: 0.6658 - val_accuracy: 0.7676 - val_loss: 0.4872
     Epoch 2/5
     196/196
                                — 11s 54ms/step - accuracy: 0.8173 - loss: 0.4257 - val_accuracy: 0.8356 - val_loss: 0.3813
     Enoch 3/5
     196/196
                                - 19s 48ms/step - accuracy: 0.8623 - loss: 0.3308 - val accuracy: 0.8637 - val loss: 0.3256
     Epoch 4/5
     196/196 -
                                - 9s 48ms/step - accuracy: 0.8834 - loss: 0.2804 - val_accuracy: 0.8665 - val_loss: 0.3170
     Epoch 5/5
     196/196 -
                                — 10s 51ms/step - accuracy: 0.9002 - loss: 0.2591 - val_accuracy: 0.8704 - val_loss: 0.3118
score = rnn.evaluate(X_test,Y_test)
→ 782/782 -
                            ---- 8s 10ms/step - accuracy: 0.8705 - loss: 0.3121
plt.figure()
plt.plot(history.history["accuracy"],label="Train");
plt.plot(history.history["val_accuracy"],label="Test");
plt.title("Accuracy")
plt.ylabel("Accuracy")
plt.xlabel("Epochs")
plt.legend()
plt.show();
```





```
plt.figure()
plt.plot(history.history["loss"],label="Train");
plt.plot(history.history["val_loss"],label="Test");
plt.title("Loss")
plt.ylabel("Loss")
plt.xlabel("Epochs")
plt.legend()
plt.show();
```

